



Detecting Structural Changes and Outliers in Longitudinal Data

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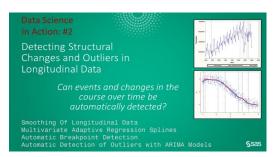


Data Science Applications and Case Studies

















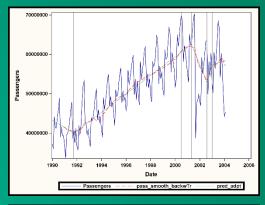


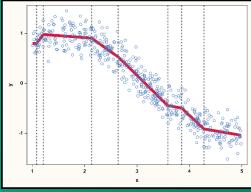
Data Science in Action: #2

Detecting Structural
Changes and Outliers in
Longitudinal Data

Can events and changes in the course over time be automatically detected?

Smoothing Of Longitudinal Data
Multivariate Adaptive Regression Splines
Automatic Breakpoint Detection
Automatic Detection of Outliers with ARIMA Models

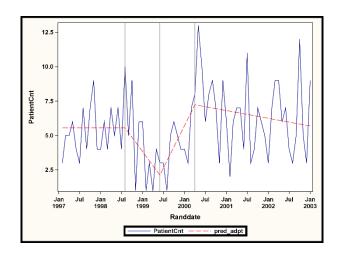




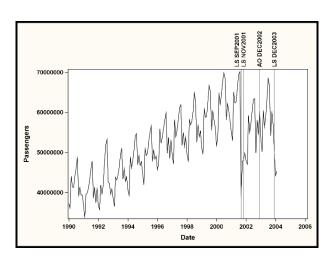


Automatically Detect Breakpoints and Outliers

Use machine learning methods to identify time points in your data where the course over time deviates from "normal" behavior.



Use multivariate regression splines to identify breakpoints over time.



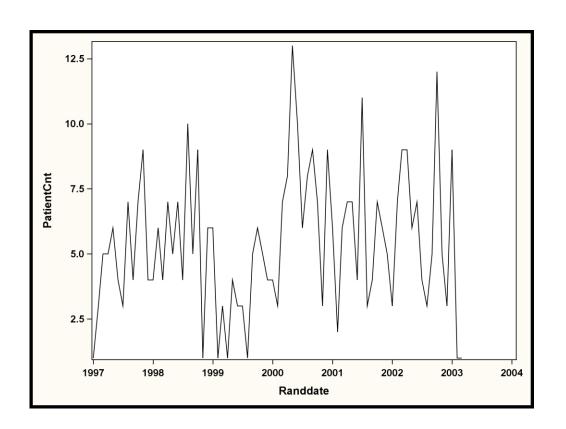
Detecting shifts and pulse events in your data with ARIMA Models.



Detecting Breakpoints



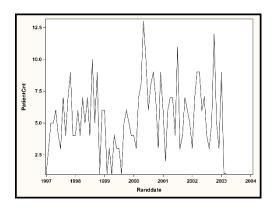
Recruitment Numbers from a Clinical Trial





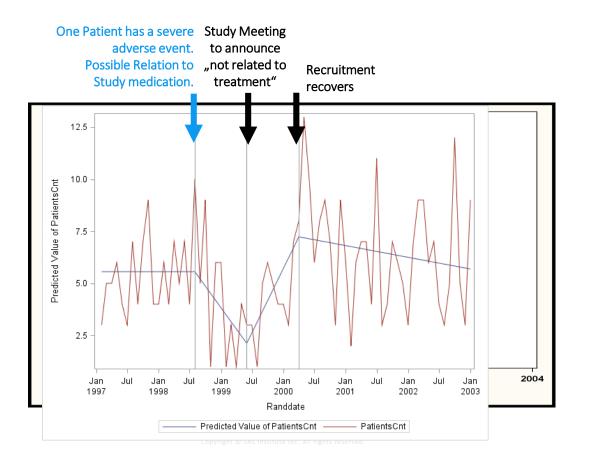
Can the breakpoints be detected automatically?

- Multivariate adaptive regression splines
 - Non-parametric regression techniques
 - Regression splines + model selection
 - ADAPTIVEREG procedure in SAS/STAT
- Breakpoints need not to be specified in advance.
- Learn from your data when a change might have taken place.



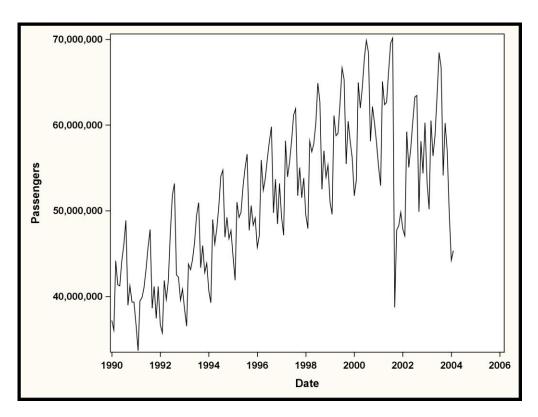


What happened in the clinical trial at certain points in time?



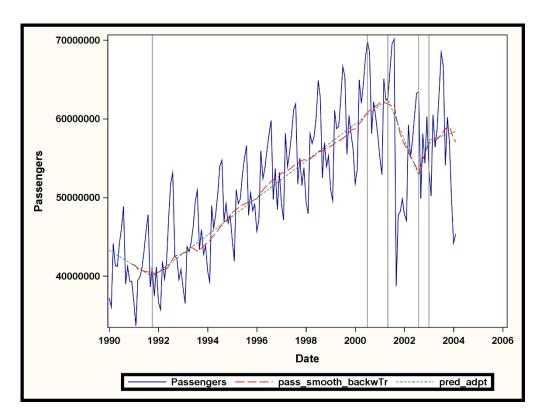


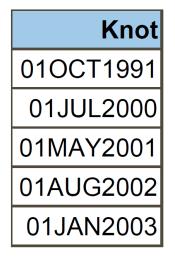
Airline Passenger Data (Monthly Sum)





Adding the automatically detected knot points to the line chart



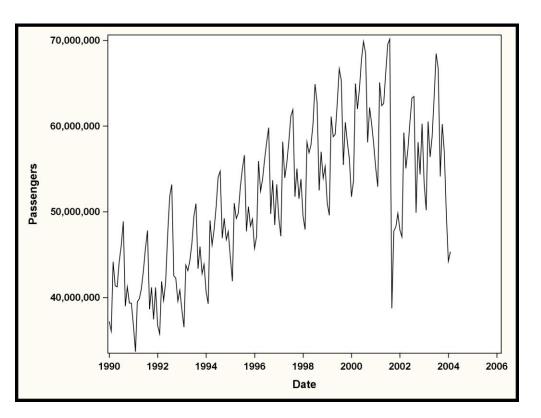




Smoothing of Longitudinal Data



Airline Passenger Data (Monthly Sum)



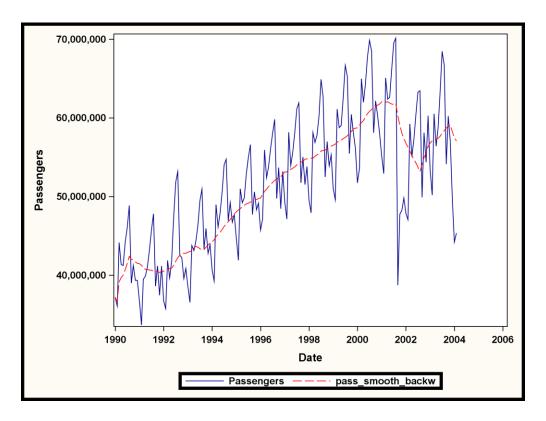


Centered and Backward Smoothing

Orientation Type	Feb 24 th	Feb 25 th	Feb 26 th	Feb 27 th	Feb 28 th	Mar 1 st	Mar 2 nd	Mar 3 rd	Mar 4 th	Mar 5 th
							Actual Date			
Centered Smoothing				-3	-2	-1	0	1	2	3
Backward Smoothing	-6	-5	-4	-3	-2	-1	0			

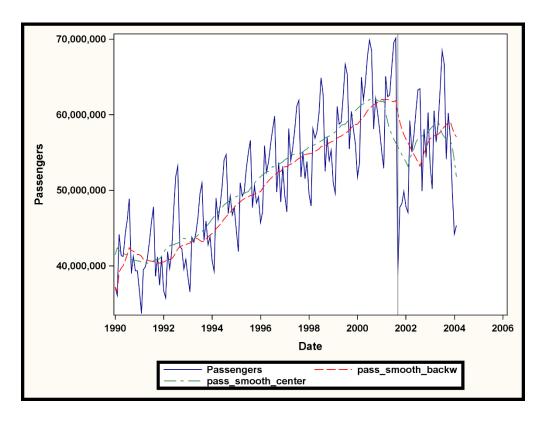


Backward Smoothing of the Airline Passenger Time Series





Backward and Centered Smoothing of the Airline Passenger Time Series



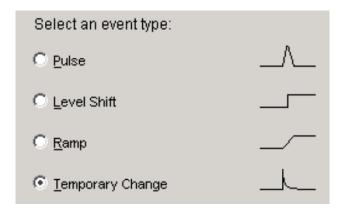


Detecting Outliers



Procedure in Detecting Outliers

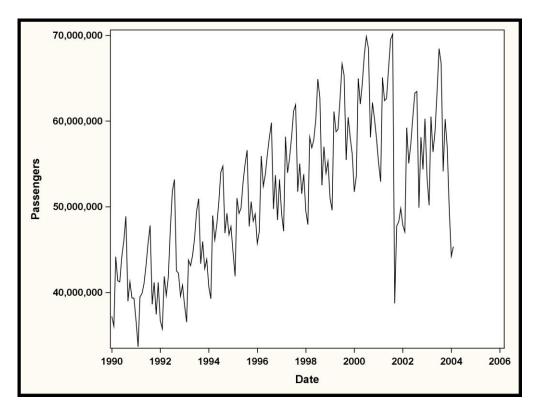
- Different types:
 - Outliers (pulse)
 - level shift
 - ramp
 - temporary change



- Fit an ARIMA model to the time series with the X13, HPFDIAGNOSE or TSMODEL procedure
- Automatically identify those points where the time series deviates from the average pattern



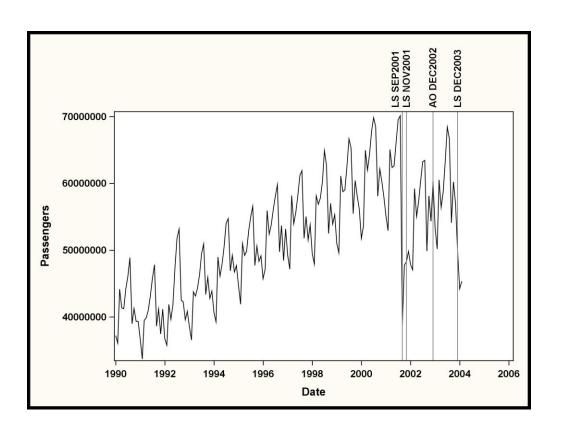
Applying Outlier Detection to the Airline Passenger Data



RegVar	Estimate				
LS SEP2001	-17,993,817.7				
LS NOV2001	5,939,640.5				
AO DEC2002	5,039,786.8				
LS DEC2003	-8,531,934.1				

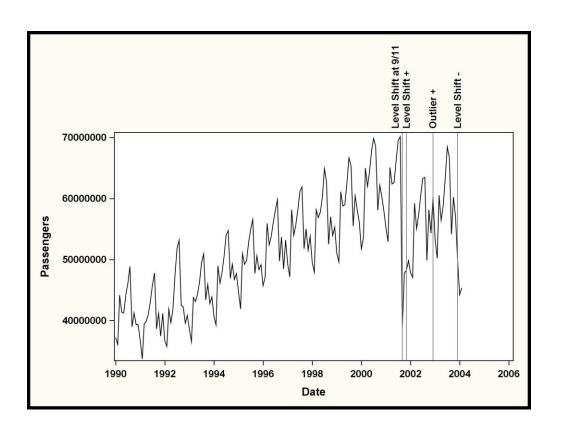


Labeling the Outliers in the Time Series Plot





Displaying individual labels





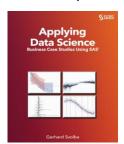
Conclusion

- Breakpoint and outlier detection can automatically detect structural changes in your time series data.
- Results are provided as lists or graphically
- Automatic insertion and labeling of the time points is key
- When smoothing time series data, be careful when selecting the smoothing method.



Analytics and Data Science is there to help you!

- Get a clearer, more objective picture of your data and your analysis subjects
- Get explicit results instead of searching the needle in the haystack
- Make your data talk to you!
- Receive findings automatically instead of manually
- Do it again! treat models as an asset and repeat your analysis



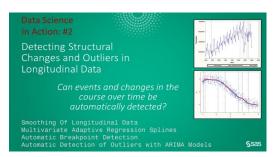


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